

A patient-specific model for convection enhanced radio-lipsome delivery

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Introduction

Re¹⁸⁶-liposome brachytherapy is a new method for the treatment of glioblastoma multiforme (GBM), currently in clinical trials^{1,2}.

- Liposomes delivered via convection-enhanced delivery (CED).
- Re¹⁸⁶ kills cancer cells through radioactive β decay.
- Slow, strong dose to delivery region
 - Re¹⁸⁶ half-life: **3.72 days**
 - β particle mean path-length: 2 mm
- Safe dose for external beam radiation therapy¹: **70 Gy**
- Safe dose for Re¹⁸⁶-liposome brachytherapy¹:

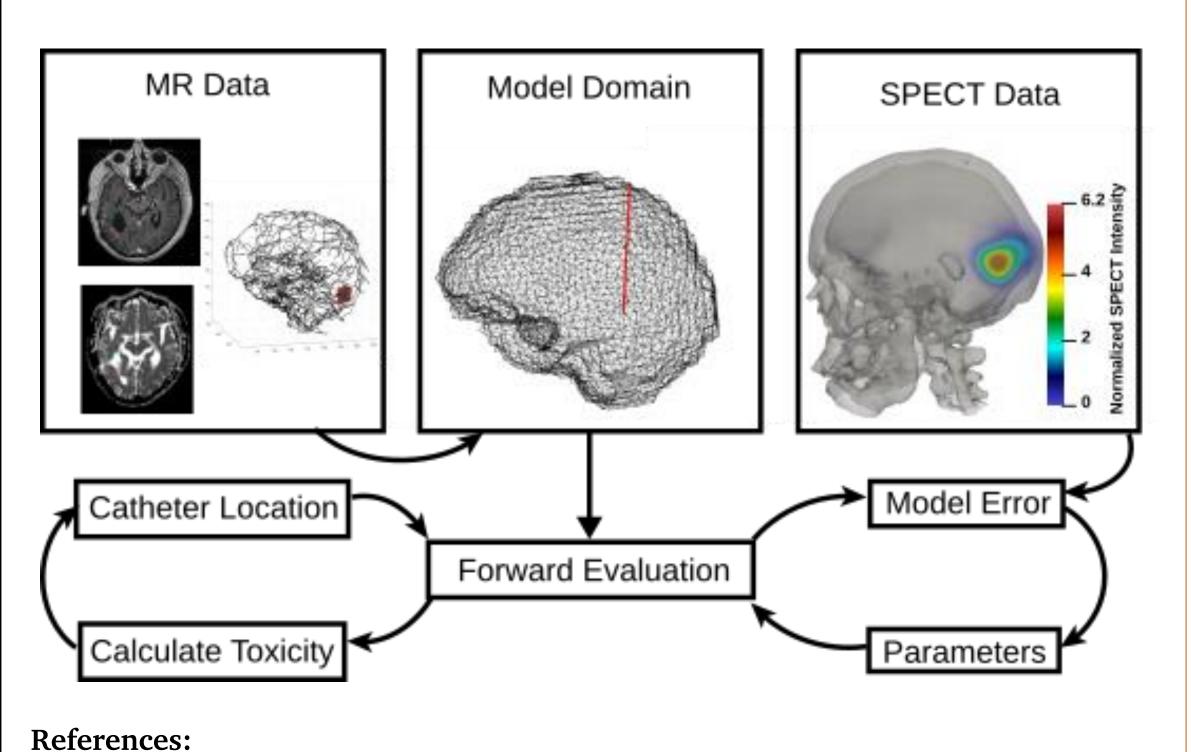
> 1800 Gy

Goal:

Develop, calibrate, and validate an advection-diffusion-reaction finite element model of nanoparticle delivery for predicting the optimal delivery location:

- Patient Specific
- Imaging informed
- Low computational resources required

Workflow



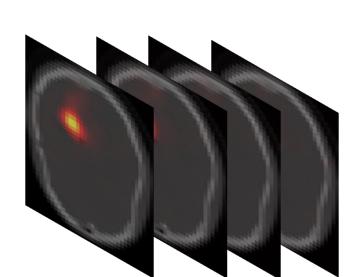
¹Phillips et al. 2012, Neuro Oncology ²Floyd et al. 2015, Neuro Oncology ³Rosenbluth et al, 2012, NeuroImage

Imaging Methods

- Use medical imaging techniques and mathematics to model liposome distribution
- T_1 , T_1 +C highlights regions of vasculature Vasculature → blood clearance
- Diffusion-weighted MRI measures the apparent diffusivity of water (ADC)
 - $ADC \rightarrow scale liposome diffusion$
- SPECT/CT measures local Re¹⁸⁶ radiation
 - Ground truth
- Reaction-Diffusion model with varying complexity

ADC T1+C

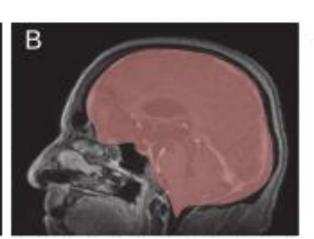
SPECT/CT

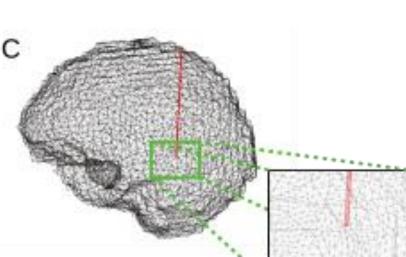


Numerical Methods

Mesh Generation







Reaction-Diffusion Equation

$$\frac{\delta C}{\delta t} = \nabla \cdot D\nabla C - R \cdot \frac{S}{V}$$

ADC-Damped Diffusion

$$D = D_0 \exp\left(-\frac{ADC}{\gamma}\right)$$

Perfusion-Informed Clearance

$$\frac{S}{V} = \frac{T_{1,post} - T_{1,pre}}{T_{1,post}}$$

- C, concentration of liposome
- D, apparent diffusivity of liposomes
- D_0 , base liposome diffusivity
- ADC, apparent diffusivity of water (diffusion-weighted MRI)
- γ , scaling factor
- R, flow rate into vasculature
- $\frac{3}{10}$, ratio of vascular surface area to vascular volume

 $T_{1,post/pre}$, T_1 before and after contrast (T_1 MRI)

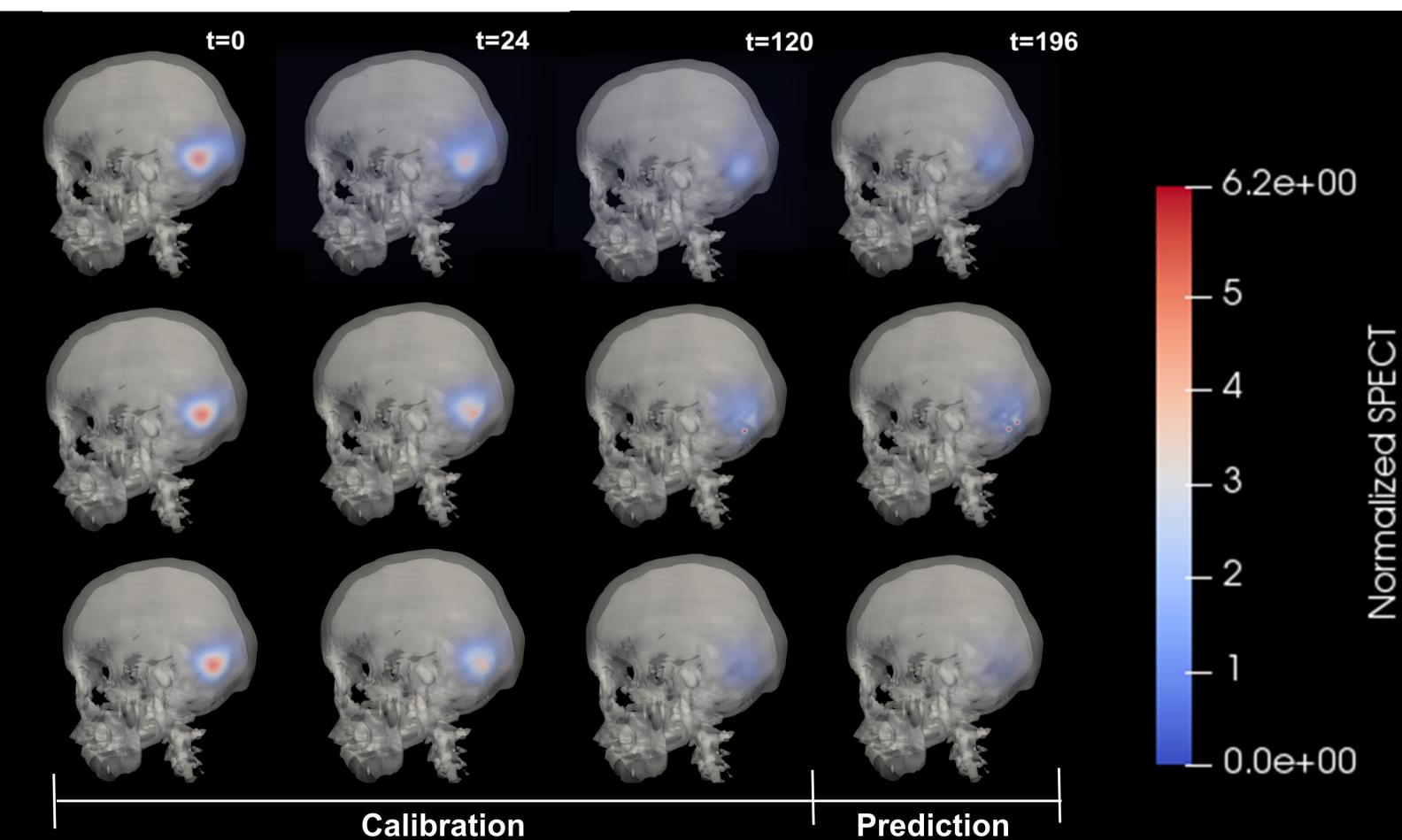
Results - Single Patient

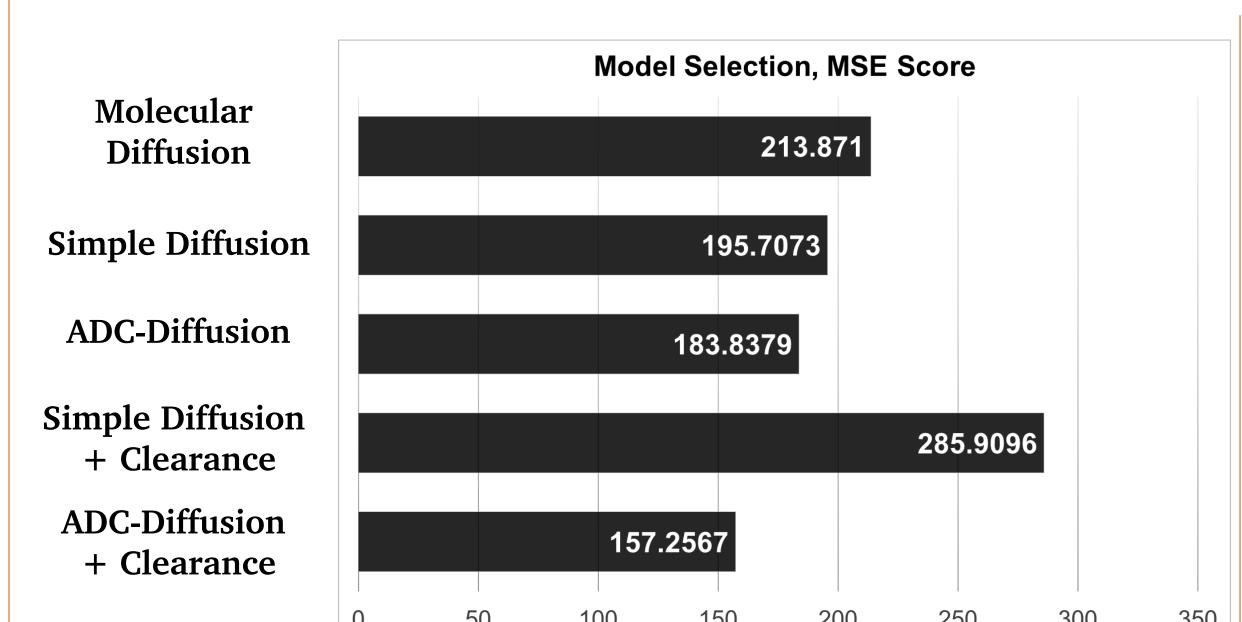
Normalized SPECT (ground truth)

> ADC-Damped Diffusion

ADC-Damped Diffusion

Perfusion-Informed Clearance





Model Calibration and Complexity Analysis

- Ground-truth: SPECT/CT
- 5 Models tested
 - Molecular Diffusion ($D=2.6E-4mm^2$)
 - Simple Diffusion (*D* calibrated)
 - ADC-Diffusion (D_0 , gamma calibrated)
 - Simple Diffusion + Clearance (D, R calibrated)
 - ADC-Diffusion + Clearance $(D_0, \gamma, R \text{ calibrated})$
- Models all calibrated through 120 hrs, and projected forward to 196 hrs
- Error measured as mean-squared error (MSE)

Future Directions + Conclusion

These results demonstrate the ability of our model to model the transport of radioactively-labeled liposomes within the brain, in comparison to software currently in use. As the ADC-Diffusion model, with clearance is the most accurate at predicting the final SPECT time point, we have selected this model for future work. Future work will focus on modeling the initial delivery of the Re¹⁸⁶-liposomes, and utilize the calibrated parameters selected for this model. Ultimately, we will use this tool to predict the localize radioactive dose absorbed within the brain, and optimize the injection location to maximizes effective dose.

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